

Random dynamical systems to analyze different ways of modeling stochastic chemostats

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1 Introduction

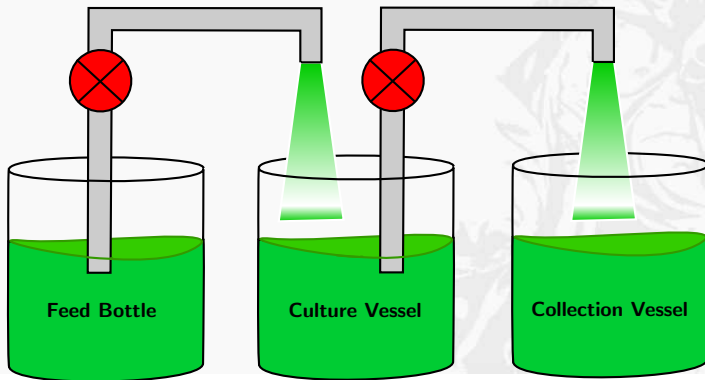
2 Our models

3 Preliminaries on stochastic processes

4 Preliminaries on the theory of random dynamical systems

What is a chemostat?

- **Laboratory device** used for growing microorganisms in a cultured environment
- **Standard assumptions** for simple chemostats:
 - **Availability** of nutrient and its supply rate are both fixed
 - **Wall growth** is not taken into account



What is a chemostat?

- However...

VERY STRONG RESTRICTIONS!

since


- real world is non-autonomous and **STOCHASTIC**
 - more realistic situation: **WALL GROWTH**
- It encourages us to study

**STOCHASTIC CHEMOSTATS
WITH WALL GROWTH**

Importance of chemostat models



- ① They play an important role in *ecological studies*
- ② They are used as models of *wastewater treatment processes*
- ③ They can be considered as starting point for other several models:
 - Problems of genetically altered organisms
 - Antibiotic production models
 - Fermentation models: **WINE**, **BEER**...!!!!!!

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The simplest chemostat model

$$\frac{dS}{dt} = (S^0 - S)D - \frac{mSx}{a + S},$$

$$\frac{dx}{dt} = -Dx + \frac{mSx}{a + S},$$

where

- $S(t)$: concentration of the nutrient
- $x(t)$: concentration of the microbial biomass
- $S^0 > 0$: input concentration
- $a > 0$: half-saturation constant
- $D > 0$: dilution rate
- $m > 0$: maximal consumption rate of the nutrient and maximal specific growth rate of microorganisms

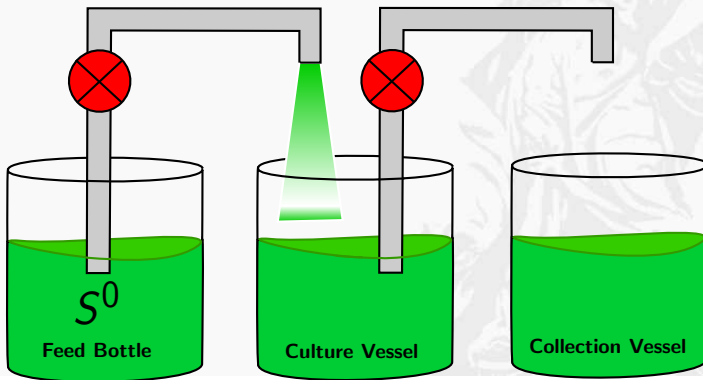
Setting up the simplest chemostat model

$$\frac{dS}{dt} = (S^0 - S)D - \frac{mSx}{a + S}$$

Equation for the nutrient

$$\frac{dx}{dt} = -Dx + \frac{mSx}{a + S}$$

Equation for the specie



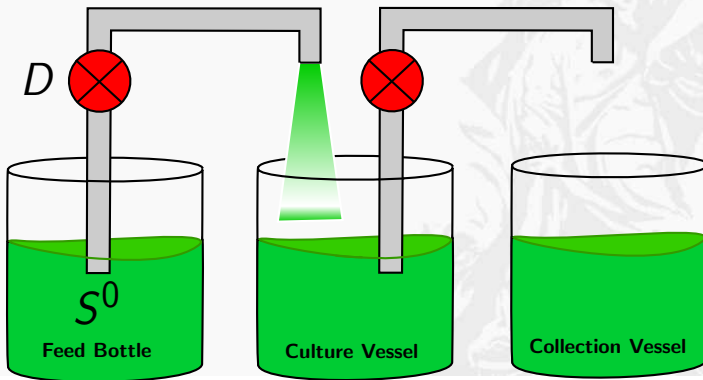
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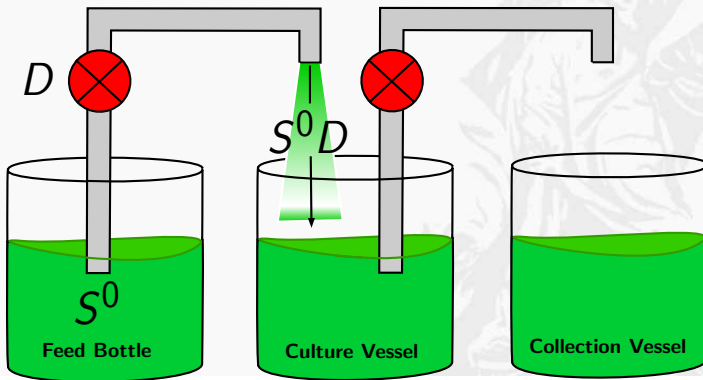
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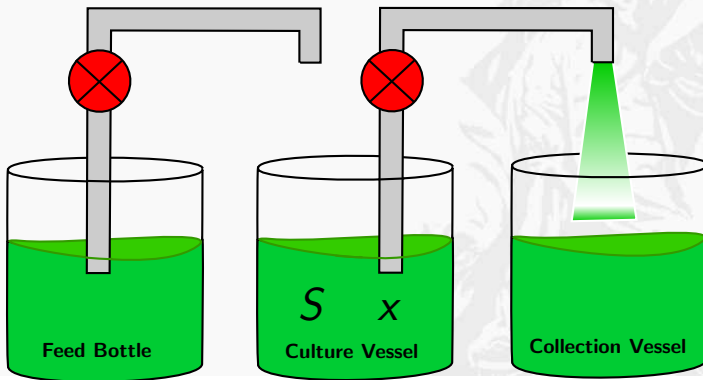
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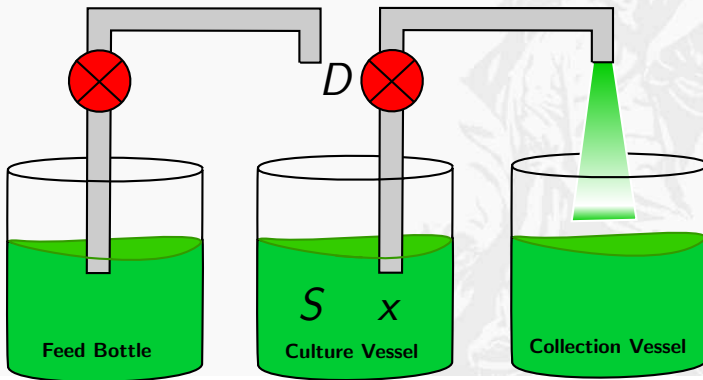
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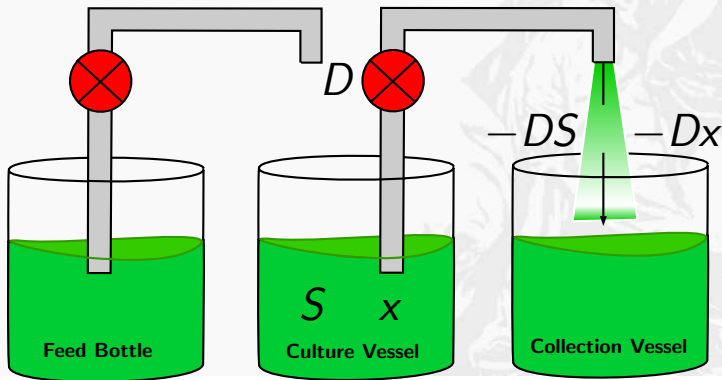
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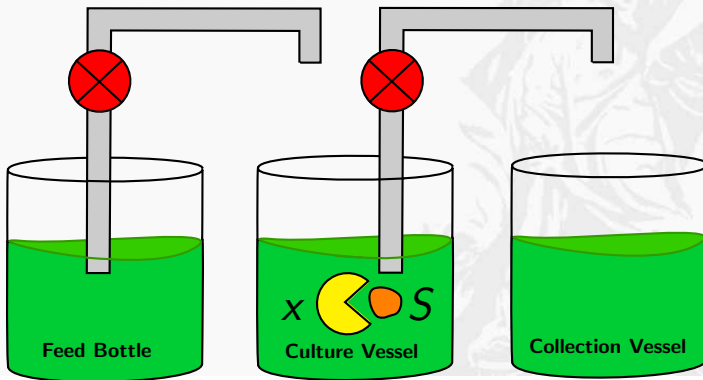
Equation for the specie



Setting up the simplest chemostat model

$$\frac{dS}{dt} = (S^0 - S)D - \frac{mSx}{a + S} \quad \text{Equation for the nutrient}$$

$$\frac{dx}{dt} = -Dx + \frac{mSx}{a + S} \quad \text{Equation for the specie}$$



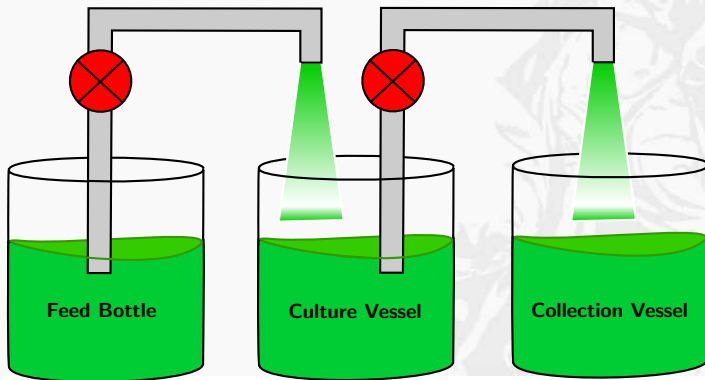
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Equation for the specie




Chemostat with wall growth

$$\frac{dS}{dt} = D(S^0 - S) - \frac{mS}{a + S}x_1 - \frac{mS}{a + S}x_2 + b\nu x_1,$$

$$\frac{dx_1}{dt} = -(\nu + D)x_1 + c\frac{S}{a + S}x_1 - r_1x_1 + r_2x_2,$$

$$\frac{dx_2}{dt} = -\nu x_2 + c\frac{S}{a + S}x_2 + r_1x_1 - r_2x_2$$

- $S(t), x_1(t), x_2(t)$: concentrations of the nutrient and the two different microorganisms
- $b \in (0, 1)$: fraction of dead biomass which is recycled
- $\nu > 0$: collective death rate coefficient
- $r_1, r_2 > 0$: rates at which the species stick on to and shear off the walls, respectively
- $0 < c \leq m$: growth rate coefficient of the consumer species

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The Wiener process

Definition

A *standard Wiener process* is a family of random variables $W(t)(\cdot) : \omega \in \Omega \mapsto W(t)(\omega) \in \mathbb{R}$, $t \geq 0$, s.t. \mathbb{P} -almost surely

- $W(0)=0$
- continuous (but NOT bounded variation) paths:
 $t \in \mathbb{R}^+ \mapsto W(t)(\omega) \in \mathbb{R}$
- independent increments: for $0 < t_1 < \dots < t_n$,
 $W(t_1), W(t_2) - W(t_1), \dots, W(t_n) - W(t_{n-1})$ are independent random variables
- stationarity: joint distribution of $\{W(t_1 + t), \dots, W(t_k + t)\}$ does NOT depend on t
- $W(t) - W(s)$, $0 \leq s \leq t$, is a Gaussian variable with mean 0 and variance $t - s$

The Wiener process

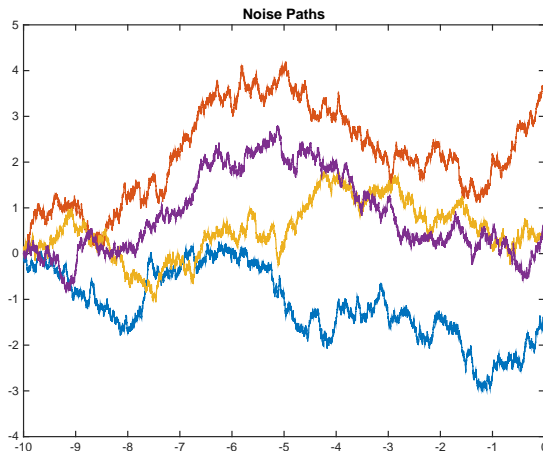


Figure: Realizations of the standard Wiener process

The Wiener process

- W : Wiener process
- *Kolmogorov's theorem* ensures that W has a continuous version, ω , whose canonical interpretation is:
 - $\Omega := \{\omega \in \mathcal{C}(\mathbb{R}, \mathbb{R}) : \omega(0) = 0\}$
 - \mathcal{F} : Borel σ -algebra on Ω
 - \mathbb{P} : Wiener measure on \mathcal{F}
- We consider the Wiener shift flow

$$\theta_t \omega(\cdot) = \omega(\cdot + t) - \omega(t), t \in \mathbb{R},$$

then $(\Omega, \mathcal{F}, \mathbb{P}, \{\theta_t\}_{t \in \mathbb{R}})$ is a *metric dynamical system*

The Ornstein-Uhlenbeck (OU) process

- The OU process on $(\Omega, \mathcal{F}, \mathbb{P}, \{\theta_t\}_{t \in \mathbb{R}})$ is defined as

$$z^*(\theta_t \omega) = - \int_{-\infty}^0 e^s \theta_t \omega(s) ds, t \in \mathbb{R}, \omega \in \Omega,$$

which solves the Langevin equation

$$dz + zdt = d\omega(t), t \in \mathbb{R} \quad (1)$$



T. Caraballo, P. E. Kloeden and B. Schmalfuß,
Exponentially stable stationary solutions for stochastic evolution
equations and their perturbation,
Applied Mathematics & Optimization, vol. **50**, no. 3, (2004)
183–207.

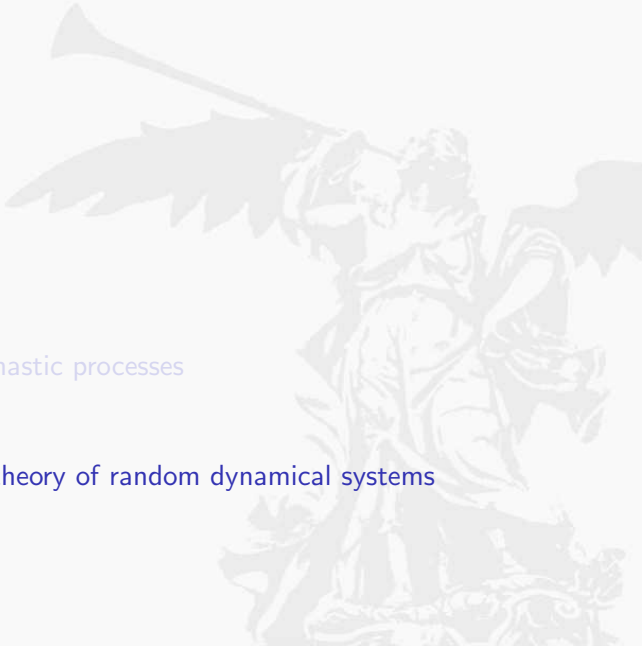
The Ornstein-Uhlenbeck (OU) process

Proposition [T. Caraballo, P. E. Kloeden and B. Schmalfuß, 2004]

There exists a θ_t -invariant set $\tilde{\Omega} \in \mathcal{F}$ of Ω of full \mathbb{P} measure such that for $\omega \in \tilde{\Omega}$, we have

- (i) the random variable $|z^*(\omega)|$ is tempered.
- (ii) the mapping $(t, \omega) \rightarrow z^*(\theta_t \omega) = - \int_{-\infty}^0 e^s \omega(t+s) ds + \omega(t)$ is a stationary solution of (1) with continuous trajectories;
- (iii) in addition, for any $\omega \in \tilde{\Omega}$:

$$\begin{aligned}\lim_{t \rightarrow \pm\infty} \frac{|z^*(\theta_t \omega)|}{t} &= 0; \\ \lim_{t \rightarrow \pm\infty} \frac{1}{t} \int_0^t z^*(\theta_s \omega) ds &= 0; \\ \lim_{t \rightarrow \pm\infty} \frac{1}{t} \int_0^t |z^*(\theta_s \omega)| ds &= \mathbb{E}[z^*] < \infty.\end{aligned}$$

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Random dynamical systems (RDSs)

$(X, \|\cdot\|_X)$ separable Banach space

Definition

An *RDS* on X consists of two ingredients: (a) a metric dynamical system $(\Omega, \mathcal{F}, \mathbb{P}, \{\theta_t\}_{t \in \mathbb{R}})$, where $(\Omega, \mathcal{F}, \mathbb{P})$ is a probability space and the family of mappings $\theta_t : \Omega \rightarrow \Omega$ satisfies

- (1) $\theta_0 = \text{Id}_\Omega$,
- (2) $\theta_s \circ \theta_t = \theta_{s+t}$ for all $s, t \in \mathbb{R}$,
- (3) the mapping $(t, \omega) \mapsto \theta_t \omega$ is measurable,
- (4) the probability measure \mathbb{P} is preserved by θ_t , i.e., $\theta_t \mathbb{P} = \mathbb{P}$

and (b) a mapping $\varphi : [0, \infty) \times \Omega \times X \rightarrow X$ which is

$(\mathcal{B}[0, \infty) \times \mathcal{F} \times \mathcal{B}(X), \mathcal{B}(X))$ -measurable, such that for each $\omega \in \Omega$,

- (i) mapping $\varphi(t, \omega) : X \rightarrow X$, $x \mapsto \varphi(t, \omega)x$ is cont. for every $t \geq 0$,
- (ii) $\varphi(0, \omega)$ is the identity operator on X ,
- (iii) (cocycle property) $\varphi(t + s, \omega) = \varphi(t, \theta_s \omega) \varphi(s, \omega)$ for all $s, t \geq 0$.

Random dynamical systems (RDSs)

Definition

A *random set* K is a measurable subset of $X \times \Omega$ w.r.t. σ -algebra $\mathcal{B}(X) \times \mathcal{F}$. Moreover K will be said a closed (compact) random set if $K(\omega) = \{x : (x, \omega) \in K\}$, $\omega \in \Omega$, is closed (compact) for \mathbb{P} -a.e. $\omega \in \Omega$.

Definition

A bounded random set $K(\omega) \subset X$ is said to be *tempered with respect to* $\{\theta_t\}_{t \in \mathbb{R}}$ if for a.e. $\omega \in \Omega$,

$$\lim_{t \rightarrow \infty} e^{-\beta t} \sup_{x \in K(\theta_{-t}\omega)} \|x\|_X = 0, \quad \text{for all } \beta > 0;$$

a random variable $\omega \mapsto r(\omega) \in \mathbb{R}$ is said to be *tempered with respect to* $\{\theta_t\}_{t \in \mathbb{R}}$ if for a.e. $\omega \in \Omega$,

$$\lim_{t \rightarrow \infty} e^{-\beta t} \sup_{t \in \mathbb{R}} |r(\theta_{-t}\omega)| = 0, \quad \text{for all } \beta > 0.$$

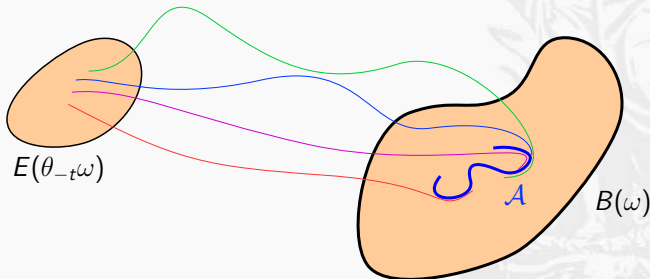
Random dynamical systems (RDSs)

$\mathcal{E}(X)$: set of all tempered random sets of X

Definition

A random set $B(\omega) \subset X$ is called a *random absorbing set* in $\mathcal{E}(X)$ if for any $E \in \mathcal{E}(X)$ and a.e. $\omega \in \Omega$, there exists $T_E(\omega) > 0$ such that

$$\varphi(t, \theta_{-t}\omega)E(\theta_{-t}\omega) \subset B(\omega), \quad \text{for all } t \geq T_E(\omega).$$



Random dynamical systems (RDSs)

Definition

Let $\{\varphi(t, \omega)\}_{t \geq 0, \omega \in \Omega}$ be an RDS over $(\Omega, \mathcal{F}, \mathbb{P}, \{\theta_t\}_{t \in \mathbb{R}})$ with state space X and let $A(\omega) (\subset X)$ be a random set. Then $\mathcal{A} = \{A(\omega)\}_{\omega \in \Omega}$ is called a *global random \mathcal{E} -attractor (or pullback \mathcal{E} -attractor)* for $\{\varphi(t, \omega)\}_{t \geq 0, \omega \in \Omega}$ if

- (i) (compactness) $A(\omega)$ is a compact set of X for any $\omega \in \Omega$;
- (ii) (invariance) for any $\omega \in \Omega$ and all $t \geq 0$, it holds

$$\varphi(t, \omega)A(\omega) = A(\theta_t \omega);$$

- (iii) (attracting property) for any $E \in \mathcal{E}(X)$ and a.e. $\omega \in \Omega$,

$$\lim_{t \rightarrow \infty} \text{dist}_X(\varphi(t, \theta_{-t} \omega)E(\theta_{-t} \omega), A(\omega)) = 0,$$

where $\text{dist}_X(G, H) = \sup_{g \in G} \inf_{h \in H} \|g - h\|_X$ is the Hausdorff semi-metric for $G, H \subseteq X$.

Random dynamical systems (RDSs)

Proposition

Let $B \in \mathcal{E}(X)$ be a closed absorbing set for the continuous RDS $\{\varphi(t, \omega)\}_{t \geq 0, \omega \in \Omega}$ that satisfies the asymptotic compactness condition for a.e. $\omega \in \Omega$, i.e., each sequence $x_n \in \varphi(t_n, \theta_{-t_n}\omega)B(\theta_{-t_n}\omega)$ has a convergent subsequence in X when $t_n \rightarrow \infty$. Then φ has a unique global random attractor $\mathcal{A} = \{A(\omega)\}_{\omega \in \Omega}$ with component subsets

$$A(\omega) = \bigcap_{\tau \geq T_B(\omega)} \overline{\bigcup_{t \geq \tau} \varphi(t, \theta_{-t}\omega)B(\theta_{-t}\omega)}.$$

Remark

The asymptotic compactness follows trivially if $X = \mathbb{R}^d$ as in the current work

Random dynamical systems (RDSs)

Lemma

Let φ_u be an RDS on X . Suppose that the mapping $\mathcal{T} : \Omega \times X \rightarrow X$ possesses the following properties: for fixed $\omega \in \Omega$, $\mathcal{T}(\omega, \cdot)$ is a homeomorphism on X , and for $x \in X$, the mappings $\mathcal{T}(\cdot, x)$, $\mathcal{T}^{-1}(\cdot, x)$ are measurable. Then the following mapping defines a (conjugated) RDS:

$$(t, \omega, x) \rightarrow \varphi_v(t, \omega)x := \mathcal{T}^{-1}(\theta_t \omega, \varphi_u(t, \omega) \mathcal{T}(\omega, x))$$



L. Arnold,

Random Dynamical Systems,
Springer-Verlag, Berlin (1998).



T. Caraballo and X. Han,

Applied Nonautonomous and Random Dynamical Systems, Applied Dynamical Systems,
Springer, 2016.



READY TO DEAL WITH STOCHASTIC CHEMOSTATS!

Modeling stochastic perturbations

- Many different ways of modeling **randomness** and **stochasticity**

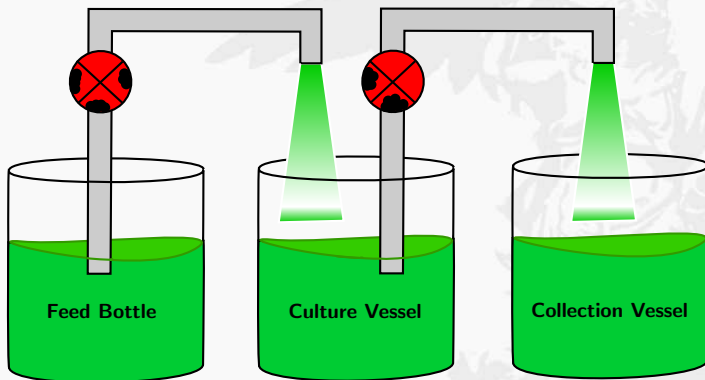
Some questions:

- What kind of stochastic perturbation can we introduce?
- How can we do it?
- Is it **realistic** from the biological point of view?
- And... Is it **tractable** from the mathematical point of view?



Perturbing the input flows

- **Biological process:** particles of dirt inside the pumps



Biologists claim that in the real life

- The flow passing through the pumps **IS NOT CONSTANT**
- The best approach: **STOCHASTIC/RANDOM FLUCTUATIONS**

How can we perturb the input flows?

We will analyze two different ways of modeling perturbed input flows in the chemostat model

- **Classical way:** by using the **standard Wiener process**

$$D \rightsquigarrow D + \alpha \dot{w}(t)$$

SEVERAL INCONVENIENTS!!

- **New idea:** by using **DIRECTLY** the **Ornstein-Uhlenbeck process**

$$D \rightsquigarrow D + \alpha z^*(\theta_t \omega)$$

EXCELLENT APPROACH TO THE REAL LIFE!!

Classical way: the Wiener process

We consider the chemostat model

$$\frac{dS}{dt} = (S^0 - S)D - \frac{mSx}{a + S},$$

$$\frac{dx}{dt} = -Dx + \frac{mSx}{a + S},$$

and introduce the following perturbation

$$D \rightsquigarrow D + \alpha \dot{\omega}(t)$$

then we obtain the following stochastic model understood in Itô's sense

$$dS = \left[(S^0 - S)D - \frac{mSx}{a + S} \right] dt + \alpha(S^0 - S)d\omega(t),$$

$$dx = \left[-Dx + \frac{mSx}{a + S} \right] dt - \alpha x d\omega(t).$$

Classical way: the Wiener process

Itô vs Stratonovich conversion

$$dX_t = a(t, X_t)dt + b(t, X_t)dW_t,$$

$$dX_t = \left(a(t, X_t) - \frac{1}{2}b(t, X_t)b'(t, X_t) \right) dt + b(t, X_t) \circ dW_t$$

Then, we obtain the following stochastic chemostat model

$$dS = \left[(S^0 - S)\bar{D} - \frac{mSx}{a + S} \right] dt + \alpha(S^0 - S) \circ d\omega(t), \quad (2)$$

$$dx = \left[-\bar{D}x + \frac{mSx}{a + S} \right] dt - \alpha x \circ d\omega(t), \quad (3)$$

in Stratonovich sense, where

$$\bar{D} := D + \frac{\alpha^2}{2}. \quad (4)$$

Classical way: the Wiener process

We define the following variable change

$$\sigma(t) = (S(t) - S^0)e^{\alpha z^*(\theta_t \omega)}, \quad (5)$$

$$\kappa(t) = x(t)e^{\alpha z^*(\theta_t \omega)}. \quad (6)$$

Then, by differentiation, we have the following **RANDOM** chemostat model given by

$$\frac{d\sigma}{dt} = -(\bar{D} + \alpha z^*)\sigma - \frac{m(S^0 + \sigma e^{-\alpha z^*(\theta_t \omega)})}{a + S^0 + \sigma e^{-\alpha z^*(\theta_t \omega)}}, \quad (7)$$

$$\frac{d\kappa}{dt} = -(\bar{D} + \alpha z^*)\kappa + \frac{m(S^0 + \sigma e^{-\alpha z^*(\theta_t \omega)})}{a + S^0 + \sigma e^{-\alpha z^*(\theta_t \omega)}}. \quad (8)$$

Classical way: the Wiener process

Now, we define $\mathcal{X} := \{(x, y) \in \mathbb{R}^2 : x \in \mathbb{R}, y \geq 0\}$

Theorem (Caraballo *et al*, 2017)

For any $\omega \in \Omega$ and any initial value $u_0 := (\sigma_0, \kappa_0) \in \mathcal{X}$, where $\sigma_0 := \sigma(0)$ and $\kappa_0 := \kappa(0)$, system (7)-(8) possesses a unique global solution $u(\cdot; 0, \omega, u_0) := (\sigma(\cdot; 0, \omega, u_0), \kappa(\cdot; 0, \omega, u_0)) \in \mathcal{C}^1([0, +\infty), \mathcal{X})$ with $u(0; 0, \omega, u_0) = u_0$. Moreover, the solution mapping generates a RDS $\varphi_u : \mathbb{R}^+ \times \Omega \times \mathcal{X} \rightarrow \mathcal{X}$ defined as

$$\varphi_u(t, \omega) u_0 := u(t; 0, \omega, u_0), \quad \text{for all } t \in \mathbb{R}^+, u_0 \in \mathcal{X}, \omega \in \Omega,$$

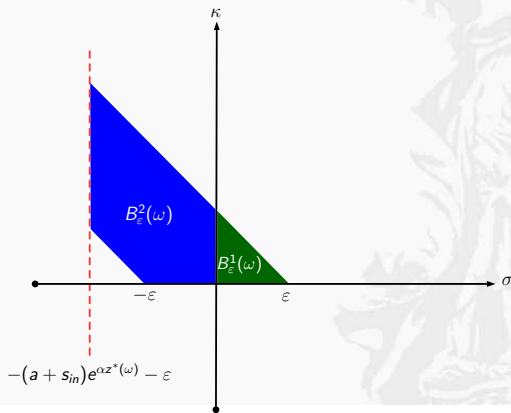
the value at time t of the solution of system (7)-(8) with initial value u_0 at time zero.

- **Key for the proof:** classical results from the theory of ODEs.
- **REMARK:** some state variable can take negative values.

Classical way: the Wiener process

Theorem (T. Caraballo *et al*, 2017)

There exists a tempered compact random absorbing set for the RDS $\{\varphi_u(t, \omega)\}_{t \geq 0, \omega \in \Omega}$.

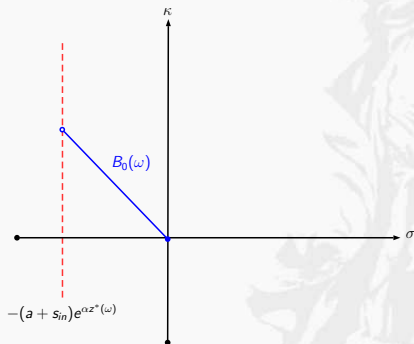


Classical way: the Wiener process

- Thanks to Proposition 25, we deduce that the RDS generated by the system (7)-(8) possesses a unique random pullback attractor given by

$$\mathcal{A} = \{A(\omega)\}_{\omega \in \Omega} \subset B_\varepsilon(\omega), \quad \text{for any } \varepsilon > 0.$$

- Thus,



$$\mathcal{A} = \{A(\omega)\}_{\omega \in \Omega} \subset B_0(\omega).$$

Classical way: the Wiener process

- Analysis of the internal structure of the random pullback attractor

Proposition (T. Caraballo *et al*, 2017)

The pullback random attractor of system (7)-(8) consists of a singleton components given by

$$\mathcal{A}(\omega) = \{(0, 0)\}$$

as long as

$$\bar{D} > \mu(S^0) \tag{9}$$

EXTINCTION OF THE MICROORGANISMS!!

Classical way: the Wiener process

- Define now a mapping $\mathcal{T} : \Omega \times \mathcal{X} \longrightarrow \mathcal{X}$ as

$$\mathcal{T}(\omega, \zeta) = \left((\zeta_1 - S^0)e^{\alpha z^*(\omega)}, \zeta_2 e^{\alpha z^*(\omega)} \right)$$

$$\mathcal{T}^{-1}(\omega, \zeta) = \left(S^0 + \zeta_1 e^{-\alpha z^*(\omega)}, \zeta_2 e^{-\alpha z^*(\omega)} \right)$$

- Denoting $v(t) = (S(t), x(t))$ and $u(t) = (\sigma(t), \kappa(t))$, since \mathcal{T} is a homeomorphism, thanks to Lemma 26 we obtain a conjugated RDS given by

$$\varphi_v(t, \omega) v_0 := v(t; 0, \omega, v_0)$$

- Hence, φ_v is an RDS for our original stochastic system (2)-(3)

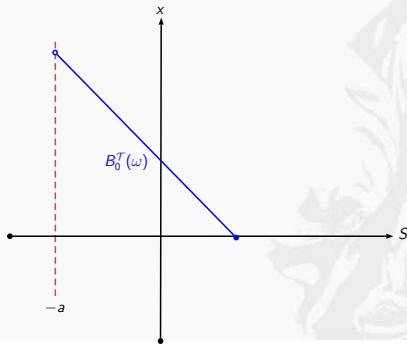
Classical way: the Wiener process

Moreover,

$$\mathcal{A} = \{A(\omega)\}_{\omega \in \Omega} \subset B_0(\omega) \implies \mathcal{A}^T = \{A^T(\omega)\}_{\omega \in \Omega} \subset B_0^T(\omega),$$

where

$$B_0^T(\omega) := \{(S, x) \in \mathcal{X} : S + x = S^0, s > -a\}$$



Recovering the stochastic chemostat

- Similarly to random case, the internal structure of the attractor consists of singleton subsets

$$A^T(\omega) = (S^0, 0)$$

as long as $\bar{D} > \mu(S^0)$

- It is not possible to ensure the persistence of the microorganism otherwise even though our simulations show that we can get the persistence for several values of the parameters

Classical way: the Wiener process

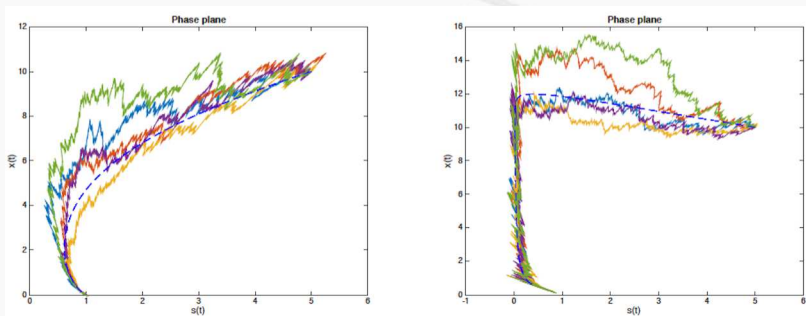


Figure: Value of parameters: $D = 3.5$, $S^0 = 1$, $a = 0.8$, $m = 1.5$ and $\alpha = 0.5$ (left), $D = 2$, $S^0 = 1$, $a = 0.6$, $m = 5$ and $\alpha = 0.5$ (right)

- Some state variables can take negative values
- The Wiener process is not a very good choice in our case

New idea: the Ornstein-Uhlenbeck process

We consider the chemostat model

$$\frac{dS}{dt} = (S^0 - S)D - \frac{mSx}{a + S},$$

$$\frac{dx}{dt} = -Dx + \frac{mSx}{a + S},$$

and introduce the following perturbation

$$D \rightsquigarrow D + \alpha z^*(\theta_{-t}\omega)$$

then we obtain the following **RANDOM** model

$$\frac{dS}{dt} = -[D + \alpha z^*(\theta_t\omega)]S - \frac{mSx}{a + S} + S^0D + \alpha S^0 z^*(\theta_t\omega), \quad (10)$$

$$\frac{dx}{dt} = -[D + \alpha z^*(\theta_t\omega)]x + \frac{mSx}{a + S}. \quad (11)$$

New idea: the Ornstein-Uhlenbeck process

Remark

Some parameters will be considered when defining the OU process

- **Langevin equation:**

$$dz = -\beta z dt + \nu d\omega(t), \quad t \in \mathbb{R}.$$

- **OU process:**

$$z_{\beta, \nu}^*(\theta_t \omega) := -\beta \nu \int_{-\infty}^0 e^{\beta s} \theta_t \omega(s) ds, \quad t \in \mathbb{R}, \omega \in \Omega, \beta, \nu > 0.$$

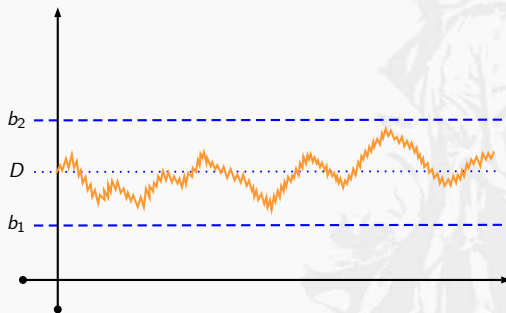
These parameters will allow us to control the noise!

New idea: the Ornstein-Uhlenbeck process

Proposition (T. Caraballo *et al*, 2017)

For any $\omega \in \Omega$,

$$\lim_{\beta \rightarrow \infty} z_{\beta, \nu}^*(\theta_t \omega) = 0, \quad \text{for all } t \in \mathbb{R}.$$



It is possible to take β large enough such that the OU is bounded

New idea: the OU process

A new framework is required

- **The new set of events:**

$$\Omega_\beta := \{\omega \in \Omega : b_1 \leq D + z_{\beta,\nu}^*(\theta_t \omega) \leq b_2, \text{ for all } t \in \mathbb{R}\},$$

- **The new σ -algebra \mathcal{F}_β :**

$$\mathcal{F}_\beta := \{A \cap \Omega_\beta, A \in \mathcal{F}\}$$

- **The new probability measure $\mathbb{P}_\beta : \mathcal{F}_\beta \rightarrow [0, 1]$:**

$$\mathbb{P}_\beta(F_\beta) := \frac{\mathbb{P}(F_\beta)}{\mathbb{P}(\Omega_\beta)}, \quad \text{for all } F_\beta \in \mathcal{F}_\beta,$$

We can make use of the theory of RDS since
 $(\Omega_\beta, \mathcal{F}_\beta, \mathbb{P}_\beta, \{\theta_t\}_{t \in \mathbb{R}}) \equiv (\Omega, \mathcal{F}, \mathbb{P}, \{\theta_t\}_{t \in \mathbb{R}})$

New idea: the OU process

Theorem (T. Caraballo *et al*, 2017)

For any $\omega \in \Omega_\beta$ and any initial value $v_0 := (S_0, x_0) \in \mathcal{X}$, where $S_0 := S(0)$ and $x_0 := x(0)$, system (10)-(11) possesses a unique global solution $v(\cdot; 0, \omega, v_0) := (S(\cdot; 0, \omega, v_0), x(\cdot; 0, \omega, v_0)) \in \mathcal{C}^1([0, +\infty), \mathcal{X})$ with $v(0; 0, \omega, v_0) = v_0$. Moreover the solution mapping generates an RDS $\varphi_v : \mathbb{R}^+ \times \Omega_\beta \times \mathcal{X} \rightarrow \mathcal{X}$ defined by

$$\varphi_v(t, \omega)v_0 := v(t; 0, \omega, v_0), \quad \text{for all } t \in \mathbb{R}^+, v_0 \in \mathcal{X}, \omega \in \Omega_\beta.$$

Theorem (T. Caraballo *et al*, 2017)

There exists a tempered compact random absorbing set $B_0(\omega) \in \mathcal{E}(\mathcal{X})$ of the RDS $\{\varphi_v(t, \omega)\}_{t \geq 0, \omega \in \Omega}$.

- The absorbing set:

$$B_0(\omega) := \{(S, x) \in \mathcal{X} : S + x = S^0\}.$$

New idea: the OU process

Proposition (T. Caraballo *et al*, 2017)

Assuming that the following condition $D > \mu(S^0)$ holds, the pullback random attractor of the chemostat model (10)-(11) is reduced to a singleton component which is given by $\mathcal{A}(\omega) = \{(S^0, 0)\}$.

In this case the microorganisms become EXTINCT!

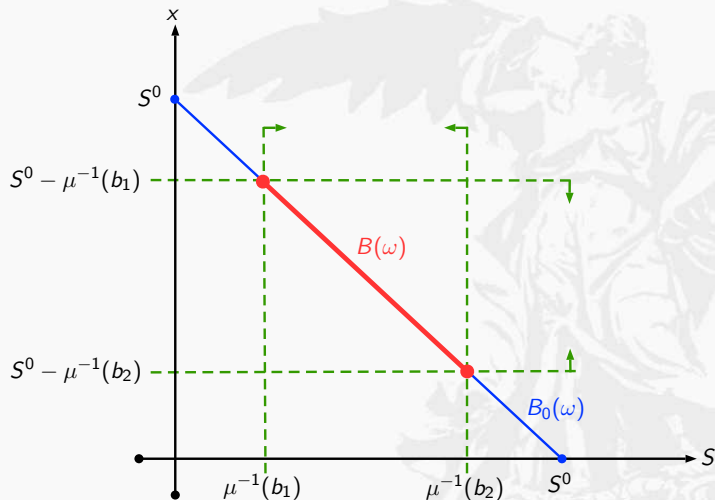
Theorem (T. Caraballo *et al*, 2017)

Assume

$$\mu^{-1}(b_2) < S^0$$

holds true. Then, there exists a **strictly positive** tempered absorbing set $B(\omega) \in \mathcal{E}(\mathcal{X})$ of the RDS $\{\varphi_v(t, \omega)\}_{t \geq 0, \omega \in \Omega}$.

New idea: the Ornstein-Uhlenbeck process



New idea: the Ornstein-Uhlenbeck process

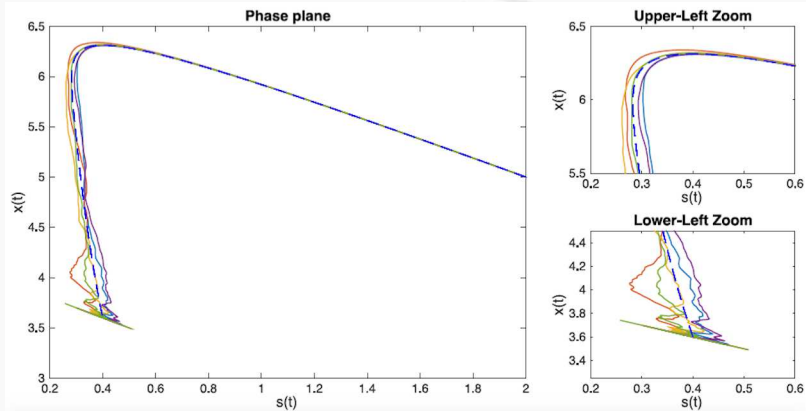


Figure: Value of parameters: $D = 2$, $S^0 = 4$, $a = 0.6$, $m = 5$, $\alpha = 0.5$, $\beta = 1$, $\nu = 0.7$, $S(0) = 2$ and $x(0) = 5$

New idea: the Ornstein-Uhlenbeck process

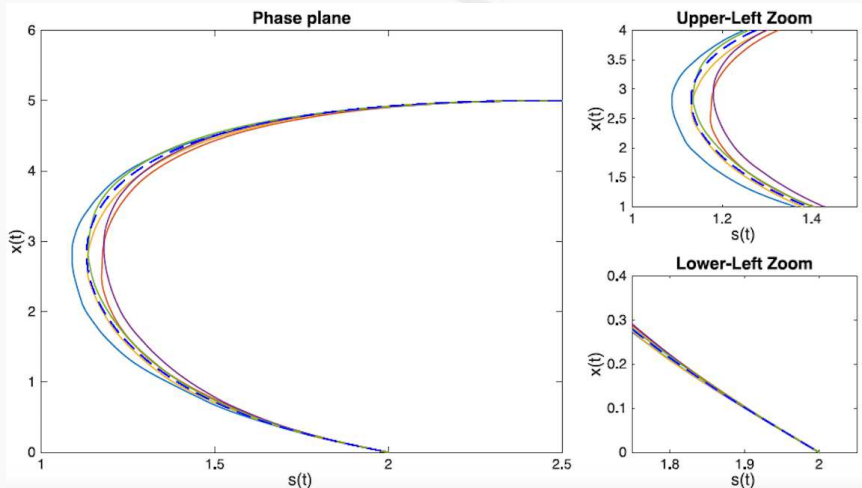


Figure: Value of parameters: $D = 3.5$, $S^0 = 2$, $a = 0.8$, $m = 0.5$, $\alpha = 0.5$, $\beta = 1$, $\nu = 0.7$, $S(0) = 2.5$ and $x(0) = 5$

Comparison: Wiener vs OU

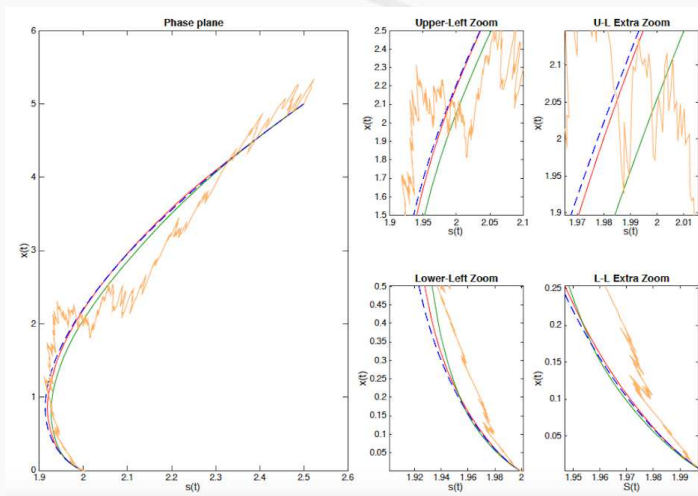


Figure: Value of parameters: $S^0 = 2$, $D = 3.5$, $a = 0.8$, $m = 0.5$, $\alpha = 0.8$, $\sigma = 0.8$, $x(0) = 5$, $S(0) = 2.5$, $\beta = 2$ (red) and $\beta = 0.5$ (green)

Comparison: Wiener vs OU

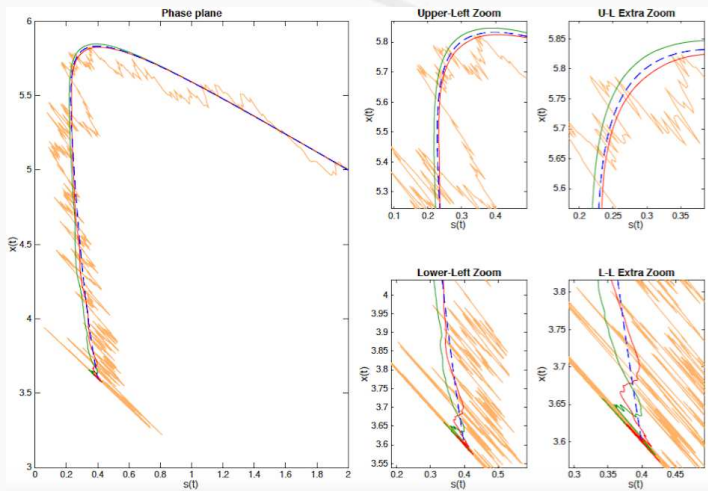


Figure: Value of parameters: $S^0 = 4$, $D = 2$, $a = 0.6$, $m = 5$, $\alpha = 0.15$, $\sigma = 0.8$, $x(0) = 5$, $S(0) = 2$, $\beta = 2$ (red) and $\beta = 0.5$ (green)

Conclusions and future works

- Classical way: $D \rightsquigarrow D + \alpha \dot{\omega}(t)$
 - Some state variables can take negative values...
 - Perturbations $D + \alpha \dot{\omega}$ are not realistic
- New way: $D \rightsquigarrow D + \alpha z_{\beta, \nu}^*(\theta_t \omega)$
 - Perturbations can be controlled by taking a suitable O-U
 - Random pullback attractor strictly contained in the positive cone
 - Hence...

**PERSISTENCE OF THE MICROORGANISM
CAN BE GUARANTEED!!!!**

Future works

- Chemostat model with wall growth.
- Generalize the new idea to perturb other models.



**¡¡¡¡Muchas gracias
por vuestra atención!!!!**

**Thank you very much
for your attention!!!!**